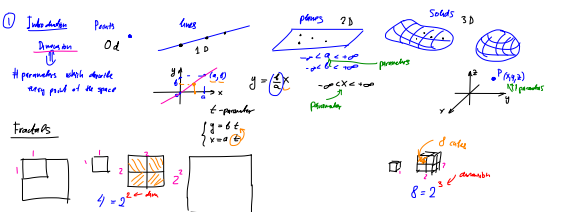


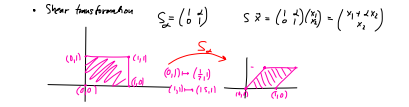
Math 22: Linear algebra and geometry
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 Lectures 1, 2, 3, 4
 add other lectures after class using T, F, R
 HW assigned every week
 Midterm on weeks 3, 4, 5
 Final on week 10
 Introduces: Algebra, Geometry, Calculus, Physics.
 Matrices: Linear algebra, diagonalization, etc.
 Algebra: Groups, Rings, Fields, Vector spaces, Eigenvalues, Eigenvectors, etc.

Vectors in \mathbb{R}^n
 $\mathbb{R}^n = \mathbb{R} \times \mathbb{R} \times \dots \times \mathbb{R}$
 $\vec{x} = (x_1, x_2, \dots, x_n)$
 Properties of operations with vectors
 commutativity: $\vec{u} + \vec{v} = \vec{v} + \vec{u}$
 group: $\vec{u} + (\vec{v} + \vec{w}) = (\vec{u} + \vec{v}) + \vec{w}$
 $\vec{u} + \vec{0} = \vec{u}$
 $\vec{u} + (-\vec{u}) = \vec{0}$
 $\alpha(\vec{u} + \vec{v}) = \alpha\vec{u} + \alpha\vec{v}$
 $(\alpha + \beta)\vec{u} = \alpha\vec{u} + \beta\vec{u}$
 $\alpha(\beta\vec{u}) = (\alpha\beta)\vec{u}$
 $1\vec{u} = \vec{u}$
 $\alpha\vec{0} = \vec{0}$
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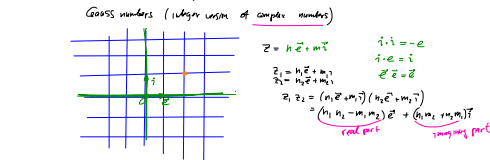
Linear Transformations
 $A: \mathbb{R}^n \rightarrow \mathbb{R}^m$
 $A_{nm} \times x_n = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} y_1 \\ \vdots \\ y_m \end{pmatrix}$
 Identity transformation: $\vec{x} \mapsto \vec{x}$
 $A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$
 Shear transformation: $S = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$
 $S \vec{x} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ x_2 \end{pmatrix}$



Linear combinations of $\vec{v}_1, \dots, \vec{v}_n$
 $\vec{v}_1, \dots, \vec{v}_n \in \mathbb{R}^n$ - vectors in \mathbb{R}^n . Then the linear combination of $\vec{v}_1, \dots, \vec{v}_n$ with coefficients c_1, \dots, c_n is $c_1\vec{v}_1 + \dots + c_n\vec{v}_n$.
 If all c_1, \dots, c_n are equal to zero then the linear combination is called **trivial**.
 If at least one of c_i is nonzero, then the linear combination is called **nontrivial**.
 Ex: $\vec{v}_1 = 5\vec{v}_2 - 10\vec{v}_3$
 If all c_1, \dots, c_n are equal to zero then the linear combination is called **trivial**.
 If at least one of c_i is nonzero, then the linear combination is called **nontrivial**.



Systems of equations according to various sets of points
 $\begin{cases} f_1(x_1, \dots, x_n) = 0 \\ f_2(x_1, \dots, x_n) = 0 \\ \vdots \\ f_m(x_1, \dots, x_n) = 0 \end{cases}$ if $f_i =$ any function
Algebraic geometry if $f_i =$ polynomial in x_1, \dots, x_n
Linear algebra if $f_i =$ linear in x_1, \dots, x_n
 $f_1(x, y) = a_1x + a_2y + a_3 = 0$
 $f_2(x, y) = a_4x + a_5y + a_6 = 0$
 where $a_1, \dots, a_6 \in \mathbb{R}$ - real or complex
 $f_1(x, y) = a_1x + a_2y + a_3 = 0$
 $f_2(x, y) = a_4x + a_5y + a_6 = 0$



Matrix notation
 $A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$
 Ex: $2x_1 + 3x_2 = 5$
 $4x_1 + 5x_2 = 7$
 $x_1 - x_2 = 1$
 $-x_1 + x_2 = 3$

Span of vectors
 $\text{Span}(\vec{v}_1, \dots, \vec{v}_n) = \{c_1\vec{v}_1 + \dots + c_n\vec{v}_n \mid c_i \in \mathbb{R}\}$
 Ex: $p=1, n=3$
 $\vec{v}_1 = (1, 1, 0)$
 $\vec{v}_2 = (0, 1, 0)$
 $\vec{v}_3 = (0, 0, 1)$
 $\text{Span}(\vec{v}_1, \vec{v}_2, \vec{v}_3) = \mathbb{R}^3$
 $\text{Span}(\vec{v}_1, \vec{v}_2) = \mathbb{R}^2$
 $\text{Span}(\vec{v}_1, \vec{v}_3) = \mathbb{R}^2$
 $\text{Span}(\vec{v}_2, \vec{v}_3) = \mathbb{R}^2$

Augmented matrix
 $\left[\begin{array}{ccc|c} 1 & -2 & 1 & 0 \\ 2 & -3 & 1 & 0 \\ 3 & 0 & -5 & 10 \end{array} \right]$
 Left hand side: $\begin{pmatrix} 1 & -2 & 1 \\ 2 & -3 & 1 \\ 3 & 0 & -5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 10 \end{pmatrix}$
 Right hand side: $\begin{pmatrix} 0 \\ 0 \\ 10 \end{pmatrix}$

Nonhomogeneous system
 $A\vec{x} = \vec{b}$
 $A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix}$
 $\vec{b} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$
 $A\vec{x} = \vec{b} \Rightarrow \begin{cases} x_1 + x_3 = 1 \\ x_2 + x_3 = 1 \\ 0 = 0 \end{cases} \Rightarrow \begin{cases} x_1 = 1 - x_3 \\ x_2 = 1 - x_3 \\ x_3 = t \end{cases} = \text{Span}(\vec{v})$

Matrix in system
 1) Multiply first equation by (-1) and add it to the third equation
 2) Multiply second equation by (-1) and add to the third equation
 3) Divide the second equation by 2, divide 3rd equation by 10
 4) Multiply third equation by 4 and add it to the second equation
 5) Multiply third equation by (-1) and add it to the first equation and add it to the first equation

Checking
 $A\vec{x} = A(\vec{v} + \vec{w}) = A\vec{v} + A\vec{w} = \vec{b} + \vec{0} = \vec{b}$

Matrix of Echelon form
 $\begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{pmatrix}$
 Echelon form: 1) All nonzero rows are above any rows with all zeros. 2) Each leading entry of a row is in a column to the right of the leading entry of the row above it. 3) All elements in a column below a leading entry are zeros.
 Reduced Echelon form: 4) The leading entry in each nonzero row is 1. 5) Each leading 1 is the only nonzero element in its column.

Linear Independence
 $\vec{v}_1, \dots, \vec{v}_n \in \mathbb{R}^n$ is linearly independent if the only linear combination of these vectors which equal to zero is trivial.
 $c_1\vec{v}_1 + \dots + c_n\vec{v}_n = \vec{0} \Rightarrow c_1 = \dots = c_n = 0$
 $\vec{v}_1, \dots, \vec{v}_n \in \mathbb{R}^n$ is linearly dependent if for $c_1\vec{v}_1 + \dots + c_n\vec{v}_n = \vec{0}$ at least one of c_i is nonzero (or there is a nontrivial linear combination which is equal to zero).

Pivot positions
 Row elimination algorithm
 A pivot position of matrix A is a location in A that corresponds to a leading 1 in the reduced Echelon form of A.

Linear Independence
 $\vec{v}_1 = (1, 0), \vec{v}_2 = (0, 1)$ in \mathbb{R}^2
 $c_1\vec{v}_1 + c_2\vec{v}_2 = \vec{0} \Rightarrow \begin{cases} c_1 = 0 \\ c_2 = 0 \end{cases}$ linearly independent
 $\vec{v}_1 = (1, 1), \vec{v}_2 = (1, 0)$ in \mathbb{R}^2
 $c_1\vec{v}_1 + c_2\vec{v}_2 = \vec{0} \Rightarrow \begin{cases} c_1 + c_2 = 0 \\ c_1 = 0 \end{cases} \Rightarrow c_2 = 0$ linearly dependent

Vectors in \mathbb{R}^2
 A vector in \mathbb{R}^2 is an ordered pair (x, y)
 $\vec{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = x_1\vec{e}_1 + x_2\vec{e}_2$
 $\vec{y} = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = y_1\vec{e}_1 + y_2\vec{e}_2$
 $\vec{x} + \vec{y} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \end{pmatrix}$
 $\alpha\vec{x} = \begin{pmatrix} \alpha x_1 \\ \alpha x_2 \end{pmatrix}$
 $\vec{x} + \vec{y} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \end{pmatrix}$

Linear Independence
 $A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$ If \vec{v}_1, \vec{v}_2 are linearly independent then $A\vec{x} = \vec{0}$ has only the trivial solution $\vec{x}_1 = \vec{x}_2 = 0$

